

A Prior Reduced Model of Dynamical Systems

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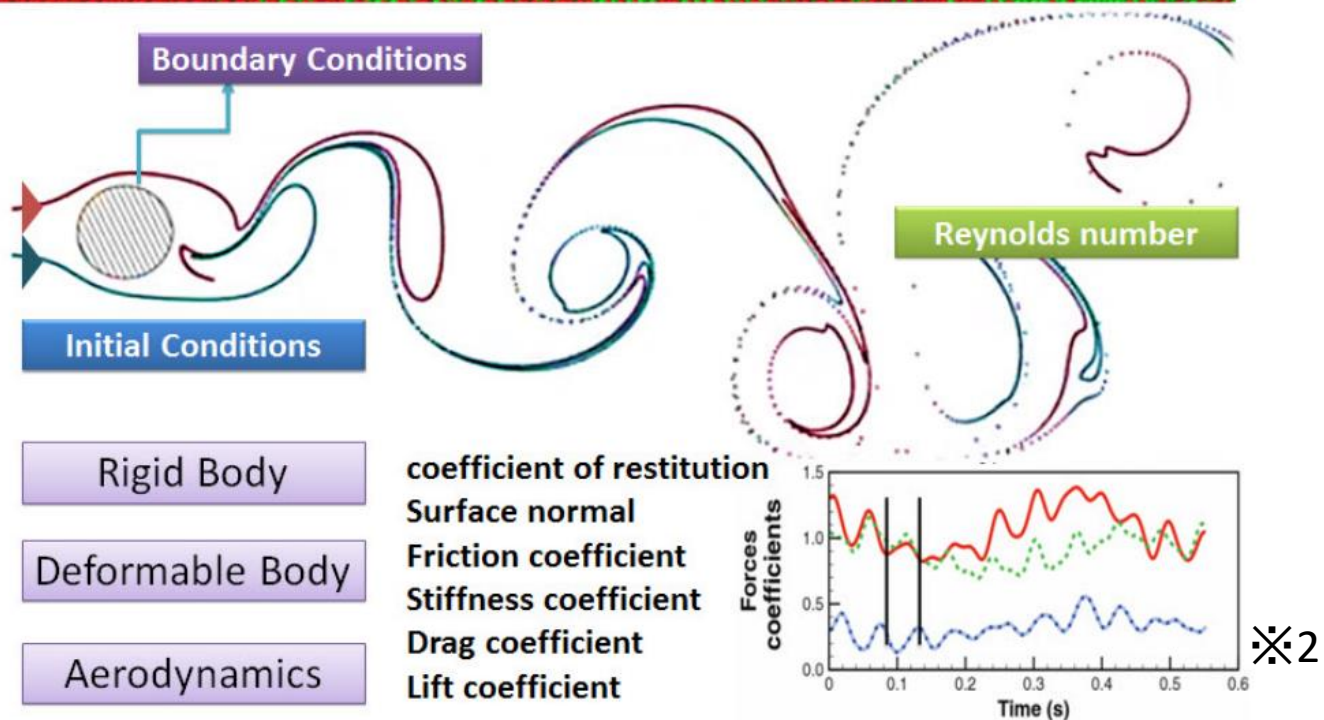
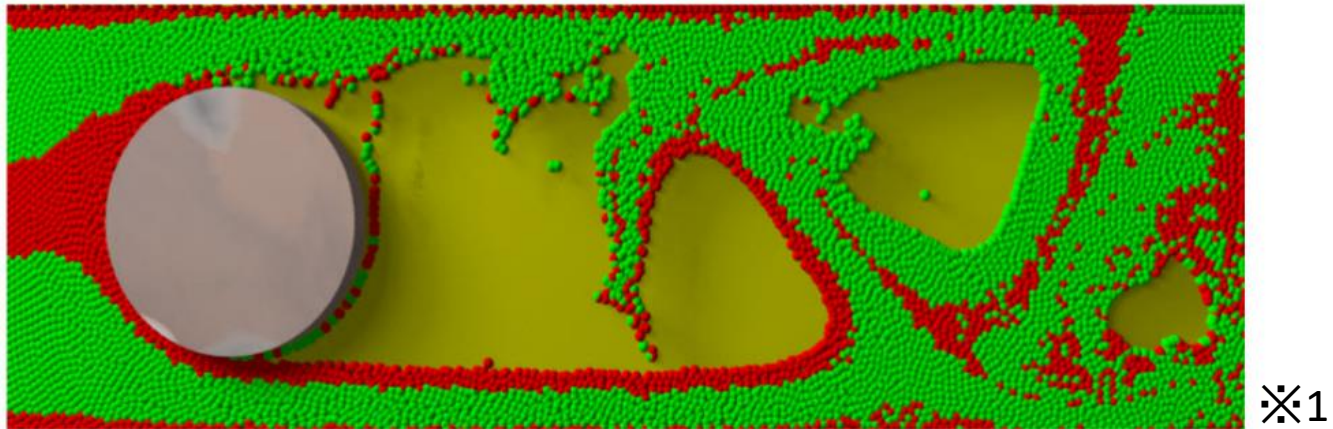


PROBLEMS

Computer Animation

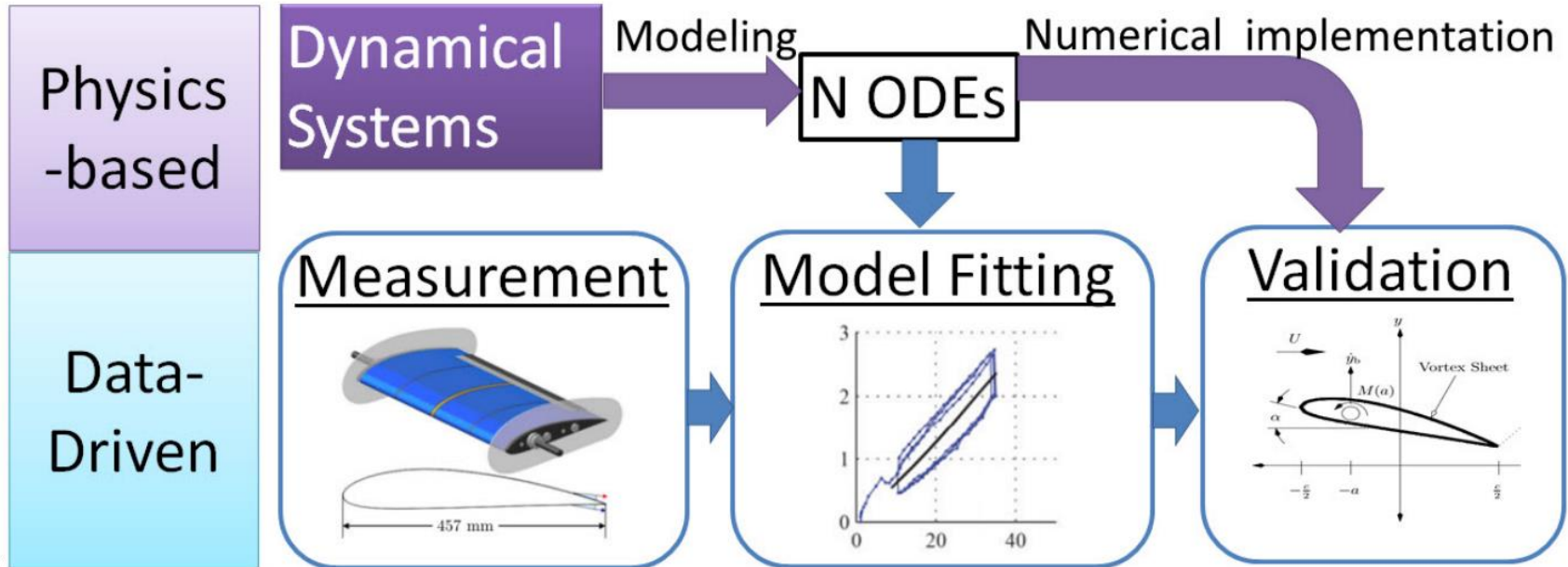
- **P1** High computation cost
- **P2** Various control parameters

As shown as

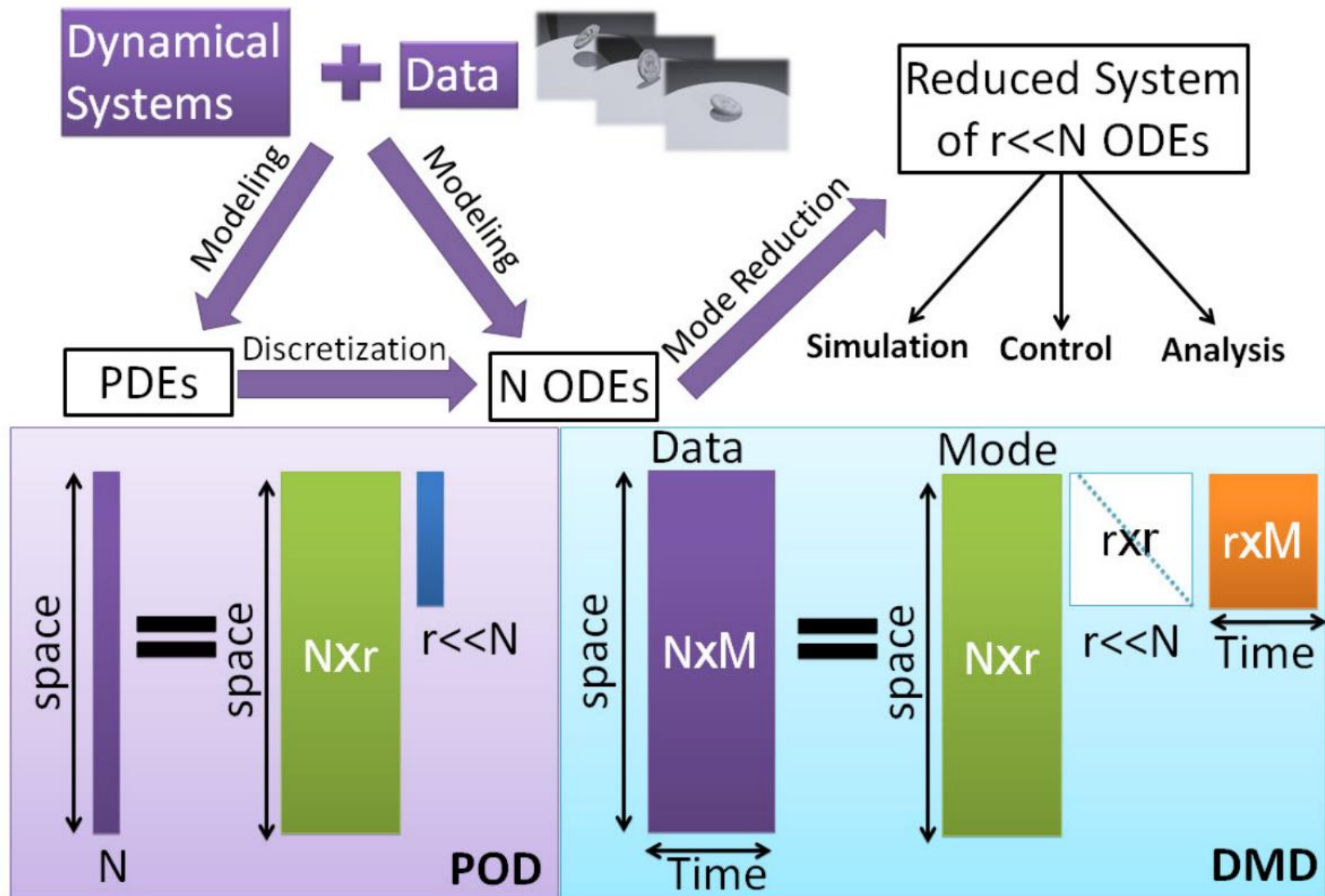


x1. He et.al, SIGGRAPH Asia 2012; 2. Vershnet et al, Physical Review E. 2013

TO Reduce computation cost



Data-driven to Model Reduction



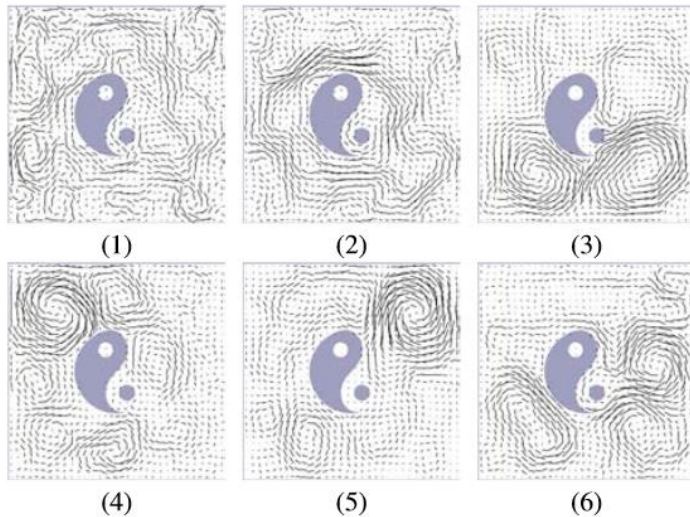
✂ POD = Proper Orthogonal Decomposition; DMD = Dynamic Mode Decomposition

P. J. Schmid (2010). Dynamic mode decomposition of numerical and experimental data. Journal of Fluid Mechanics

Model Reduction in CG

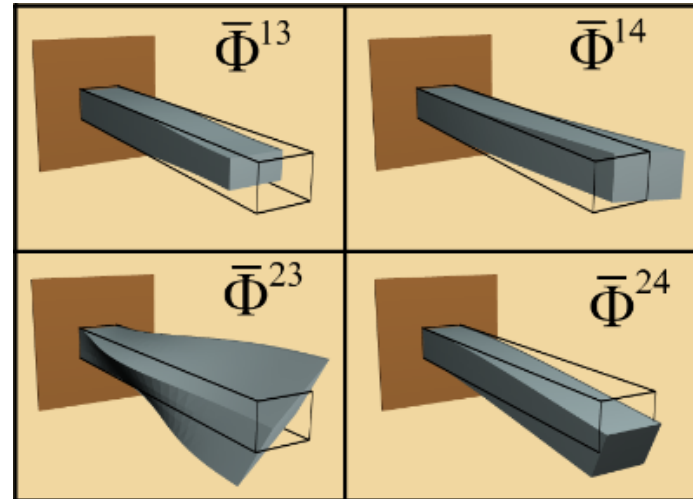
Fluid simulation

[Treuille et al., 2006]



Deformable Models

[Barbic et al., 2005]



[T. de Witt et al., 2012]



[Zhao et al., 2013]



TO find control parameters

e.g., drag/lift coefficients

$$\mathbf{F}_{a \rightarrow w} = \frac{1}{2} C_D \rho_a \pi r^2 |\mathbf{u}_{\text{rel}}| \mathbf{u}_{\text{rel}}$$

$$F_{\text{drag}}^2 = \frac{C_D}{2} \rho_g \pi \tilde{R}^2 \Delta u^2$$

$$\mathbf{f}_i^{\text{drag}} = -k_{\text{drag}} r_i^2 |\mathbf{v}_i - \mathbf{u}_i| (\mathbf{v}_i - \mathbf{u}_i)$$

$$\mathbf{f}_i^{\text{lift}} = -k_{\text{lift}} V_i (\mathbf{v}_i - \mathbf{u}_i) \times \boldsymbol{\omega}_i,$$

$$F_D = \frac{C_D}{2} \rho_L \pi r^2 \mathbf{w}_{\text{rel}} |\mathbf{w}_{\text{rel}}|$$

$$\mathbf{f}_i^{\text{drag}} = -k_{\text{drag}} r_i^2 |\mathbf{v}_{\oplus} - \mathbf{u}_{\oplus}| (\mathbf{v}_{\oplus} - \mathbf{u}_{\oplus}),$$

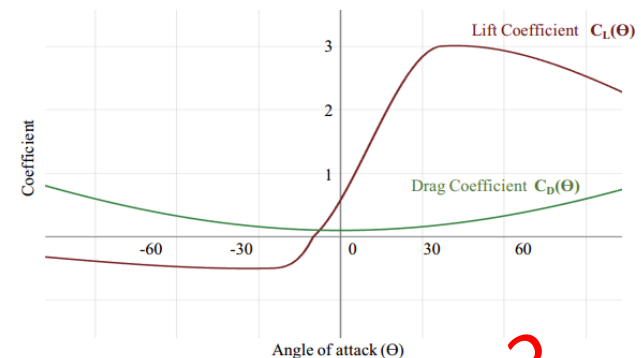
[SCA06,S09,S13,EG09,TVC12]

Constant

Reynolds
number

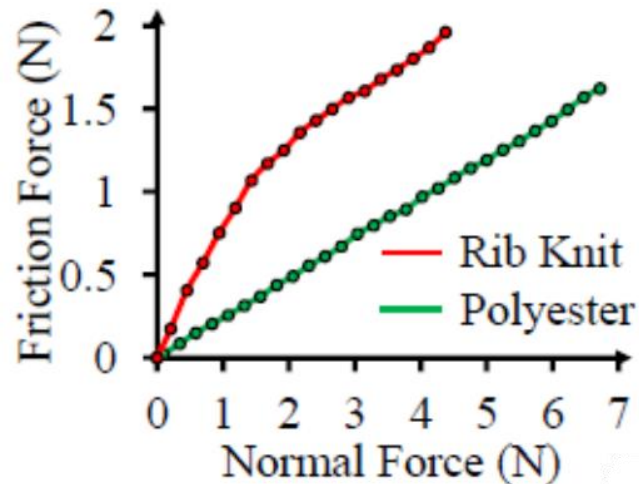
Angle of
attack

$$C_D = \frac{24}{Re^{0.72}}$$

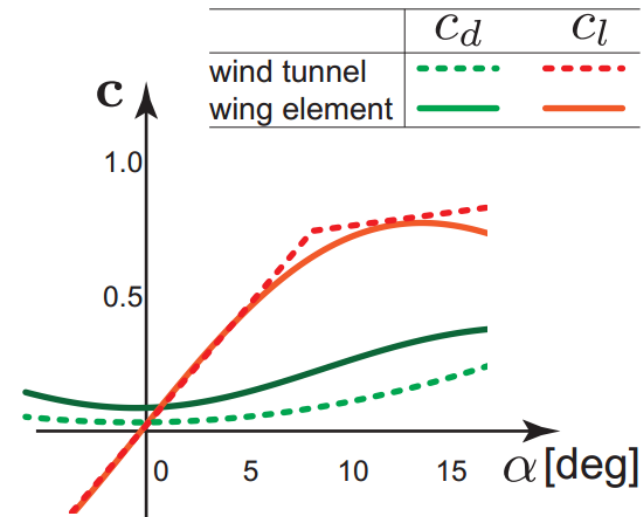


Which is proper data?

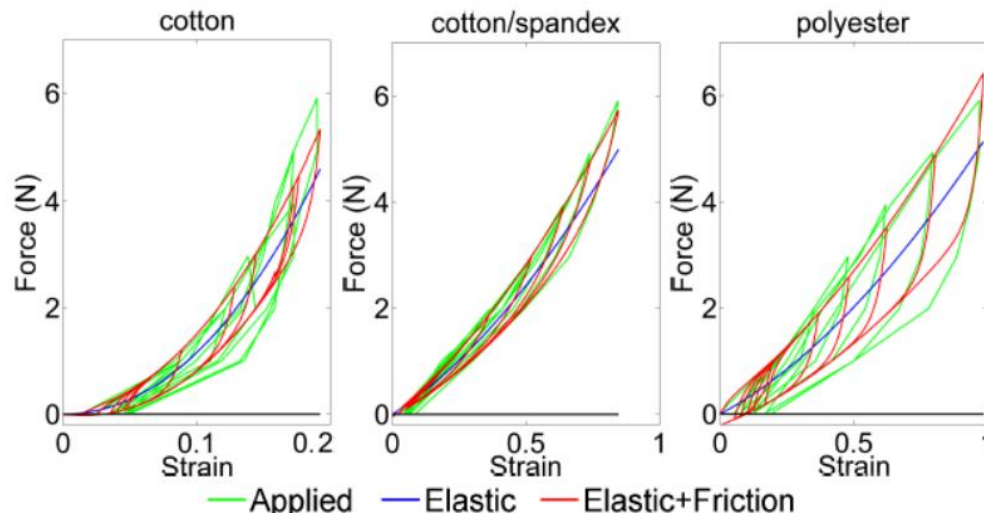
Data-driven to Parameter Estimation



Modeling Friction and Air Effects between Cloth and Deformable Bodies, S2013



Interactive Design and Optimization of Free-formed Free-flight Model Airplanes, S2014



Modeling and Estimation of Internal Friction in Cloth, SA2013

H.XIE@MEIS2014

Why others is improper?

Curse of Dimensionality

Dynamical Systems

INPUTS:

Parameter1 N nodes

Parameter2 N nodes

Parameter3 N nodes

...



$$N \times N \times N \cdots$$



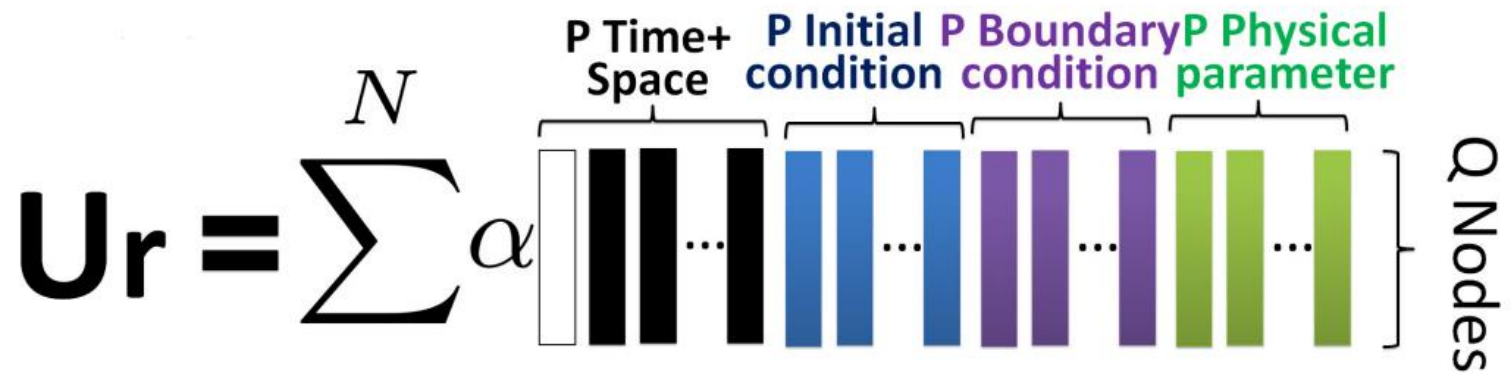
A Prior Reduced Model of Dynamical Systems

SEPARATED REPRESENTATION

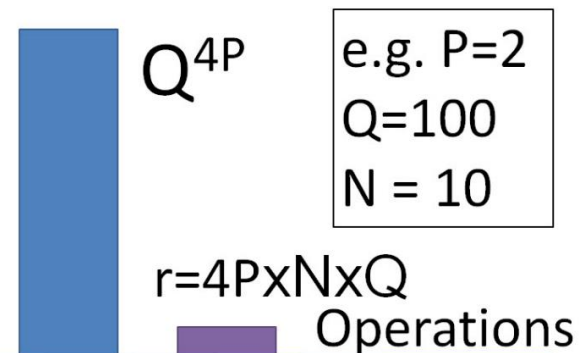
Separated Representation

Dynamical system: $D(U) = G(U, P)$

$$\text{State: } U(x_1, x_2, \dots, x_d) = \sum_{i=1}^N \alpha_i \prod_{j=1}^d U_i^j(x_j)$$



- Model-based (no precomputed snapshots)
- Reduced Model



Reduction Solver

- Determine prior unknown functions $U_i^j(x_j)$

step n .

Test functions:

$$\alpha_n = 1$$

$$U = \sum_{i=1}^{n-1} \alpha_i \prod_{j=1}^d U_i^j(x_j) + \prod_{j=1}^d U_n^j(x_j)$$

known

Weak form:

$$\langle D(U), U_n^j \rangle_{\Omega_j} = \langle G, U_n^j \rangle_{\Omega_j}$$



Iteration

Enrichment step:

Alternating directions
fixed-point method

p -th iteration

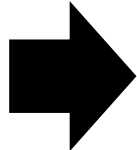
$$(u_{p-1}^2, u_{p-1}^3, \dots, u_{p-1}^d) \Rightarrow u_p^1$$

...

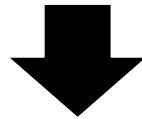
$$(u_p^1, \dots, u_p^{k-1}, u_{p-1}^{k+1}, \dots, u_{p-1}^d) \Rightarrow u_p^k$$

Check convergence

Reduction Solver

 Determine α_n

Projection step: $\langle D(U), \prod_{j=1}^d U_i^j(x_j) \rangle = \langle G, \prod_{j=1}^d U_i^j(x_j) \rangle$



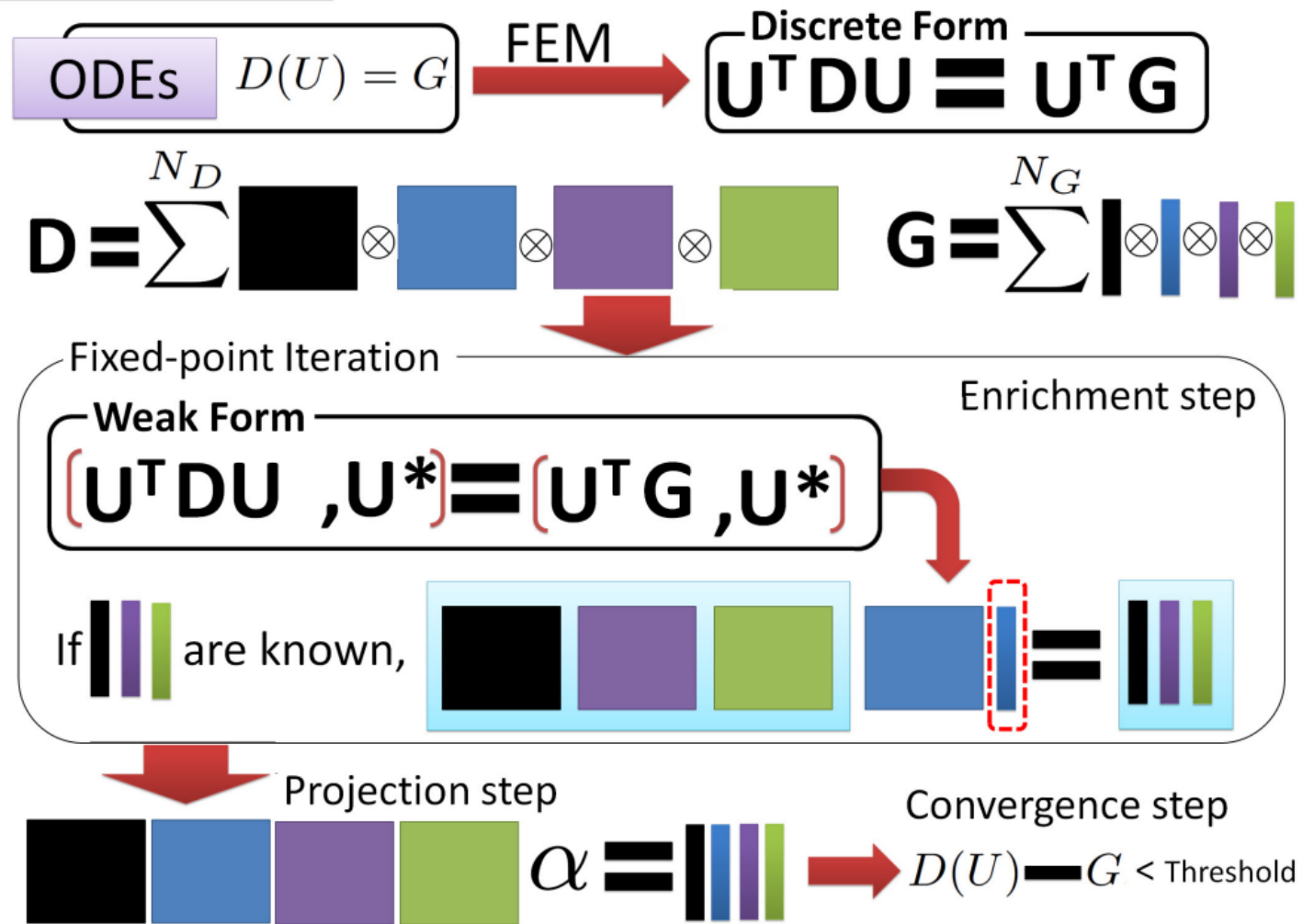
Check convergence:

Residual term

$$R^n = D\left(\sum_{i=1}^{n-1} \alpha_i \prod_{j=1}^d U_i^j(x_j) + \prod_{j=1}^d U_n^j(x_j)\right) - G < \epsilon$$

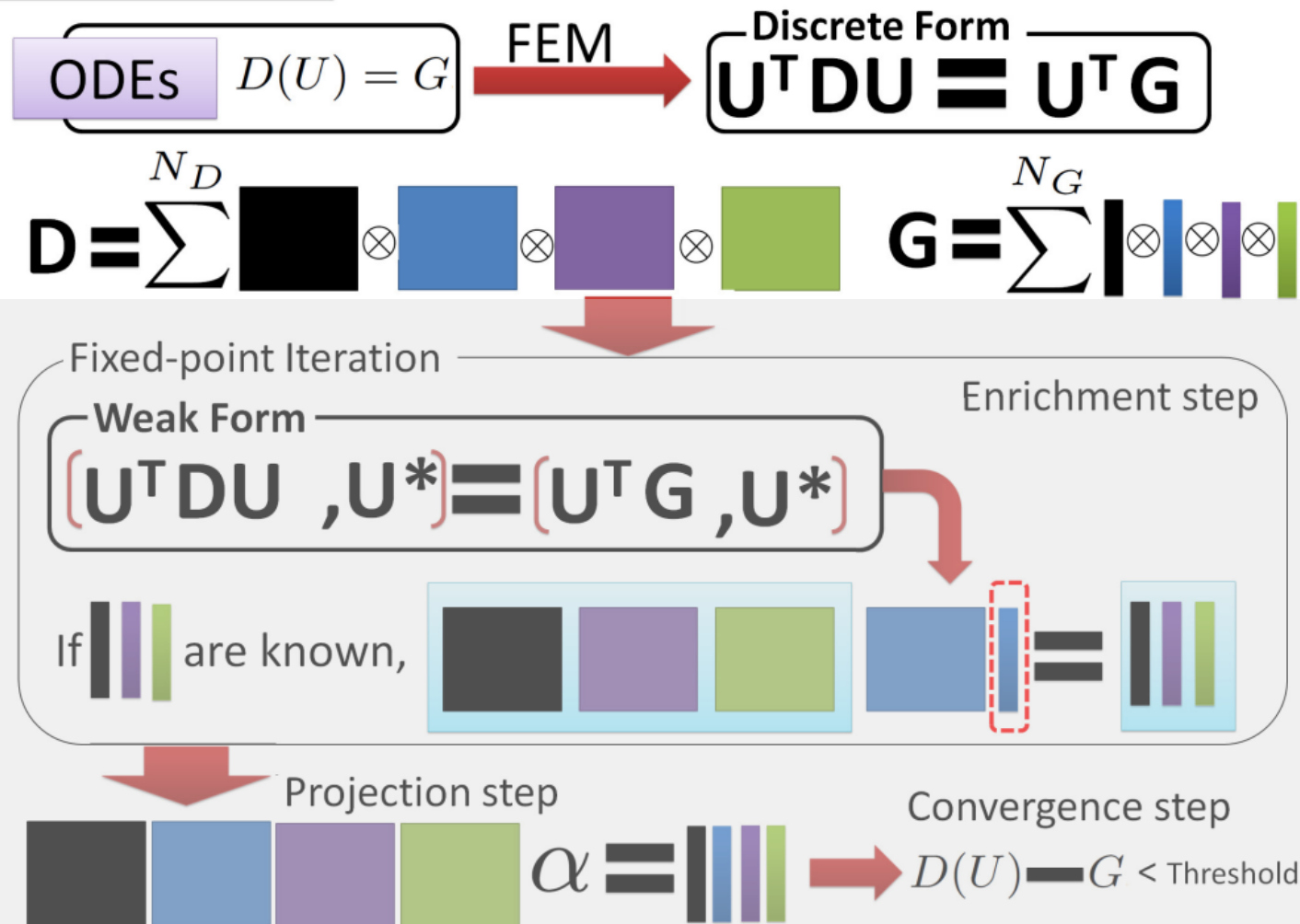
A practical framework

Model Reduction



Initialization

Model Reduction



Discrete Formulation

An example

$$\frac{du}{dt} + ku = 0$$

$$u(t, k) = \sum_{i=1}^N T_i(t) K_i(k) \\ \text{in } \Omega_t \times \Omega_k$$

Weak form: $\langle D(U), U_n^j \rangle_{\Omega_j} = \langle G, U_n^j \rangle_{\Omega_j}$



$$\langle \frac{dT_n}{dt}, T_n \rangle \langle K_n, K_n \rangle + \langle T_n, T_n \rangle \langle k K_n, K_n \rangle = - \sum_{i=1}^{n-1} (\langle \frac{dT_i}{dt}, T_n \rangle \langle K_i, K_n \rangle + \langle T_i(t), T_n \rangle \langle k K_i(k), K_n \rangle)$$



$$T^T P T \cdot K^T M_k K + T^T M_t T \cdot K^T N K = \\ - \sum_{i=1}^{n-1} (T_i^T P T \cdot K_i^T M_k K + T_i M_t T \cdot K_i^T N K)$$



$$\begin{cases} P_{ij} = \int_{\Omega_t} \frac{dT_i}{dt} N_j dt \\ N_{ij} = \int_{\Omega_k} N_i k N_j dk \\ M_t = \int_{\Omega_t} N_i N_j dt \\ M_k = \int_{\Omega_k} N_i N_j dk \end{cases}$$

shape functions ¹⁵

Algebraic Formulation

$$T^T P T \cdot K^T M_k K + T^T M_t T \cdot K^T N K = - \sum_{i=1}^{n-1} (T_i^T P T \cdot K_i^T M_k K + T_i M_t T \cdot K_i^T N K)$$



$$D(U) = G(U, P) \Rightarrow \mathbf{U}^T \mathbf{D} \mathbf{U} = \mathbf{U}^T \mathbf{G}$$

$$U = \sum_{i=1}^N \alpha_i U_1^i \otimes U_2^i \dots \otimes U_d^i$$

$\boxed{N_j}$

$$D = \sum_{i=1}^{N_D} D_1^i \otimes D_2^i \dots \otimes D_d^i$$

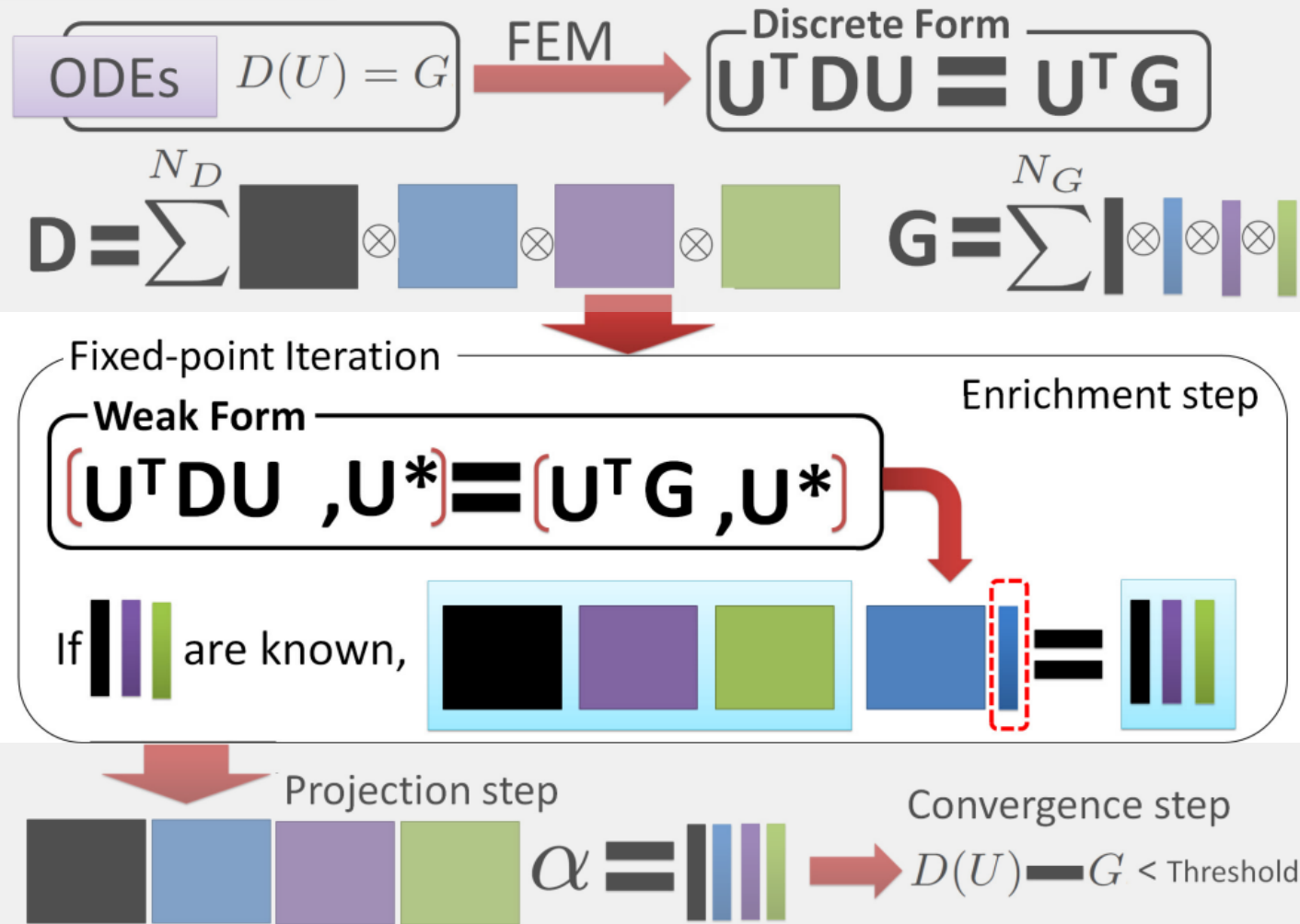
$\boxed{N_j \times N_j}$

$$G = \sum_{i=1}^{N_G} G_1^i \otimes G_2^i \dots \otimes G_d^i$$

$\boxed{N_j}$

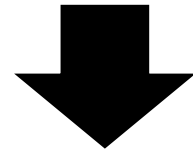
Enrichment Step

Model Reduction



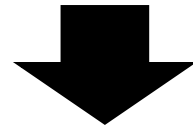
Enrichment step

Test functions: $U_n = \alpha_n R_1 \otimes R_2 \dots \otimes R_d$
 $\alpha_n = 1$



$$\mathbf{U}^T \mathbf{D} \mathbf{U} = \mathbf{U}^T \mathbf{G}$$

$$\sum_{i=1}^{N_D} D_1^i R_1 \otimes D_2^i R_2 \dots \otimes D_d^i R_d = G - \sum_{i=1}^{N_D} \sum_{k=1}^{n-1} \alpha_i D_1^i U_1^k \otimes D_2^i U_2^k \dots \otimes D_d^i U_d^k$$



- Fixed-point method

$$\boxed{p\text{-th iteration}} \quad (R_1^p, \dots, R_{j-1}^p R_{j+1}^{p-1}, \dots, R_d^{p-1}) \rightarrow R_j^p$$

$$E R_j = \sum_{i=1}^{N_G} \left(\prod_{k=1, k \neq j}^d R_k^T G_k^i \right) G_j^i - \sum_{i=1}^{N_D} \sum_{k=1}^{n-1} \left(\prod_{m=1, m \neq j}^d R_m^T D_m^i U_m^i \right) \alpha^k D_j^i U_j^k \quad \Leftarrow \text{Weak form}$$

$$\text{where } E = \sum_{i=1}^{N_D} \left(\prod_{k=1, k \neq j}^d R_k^T D_k^i R_k \right) D_j^i$$

Enrichment step

Check convergence:

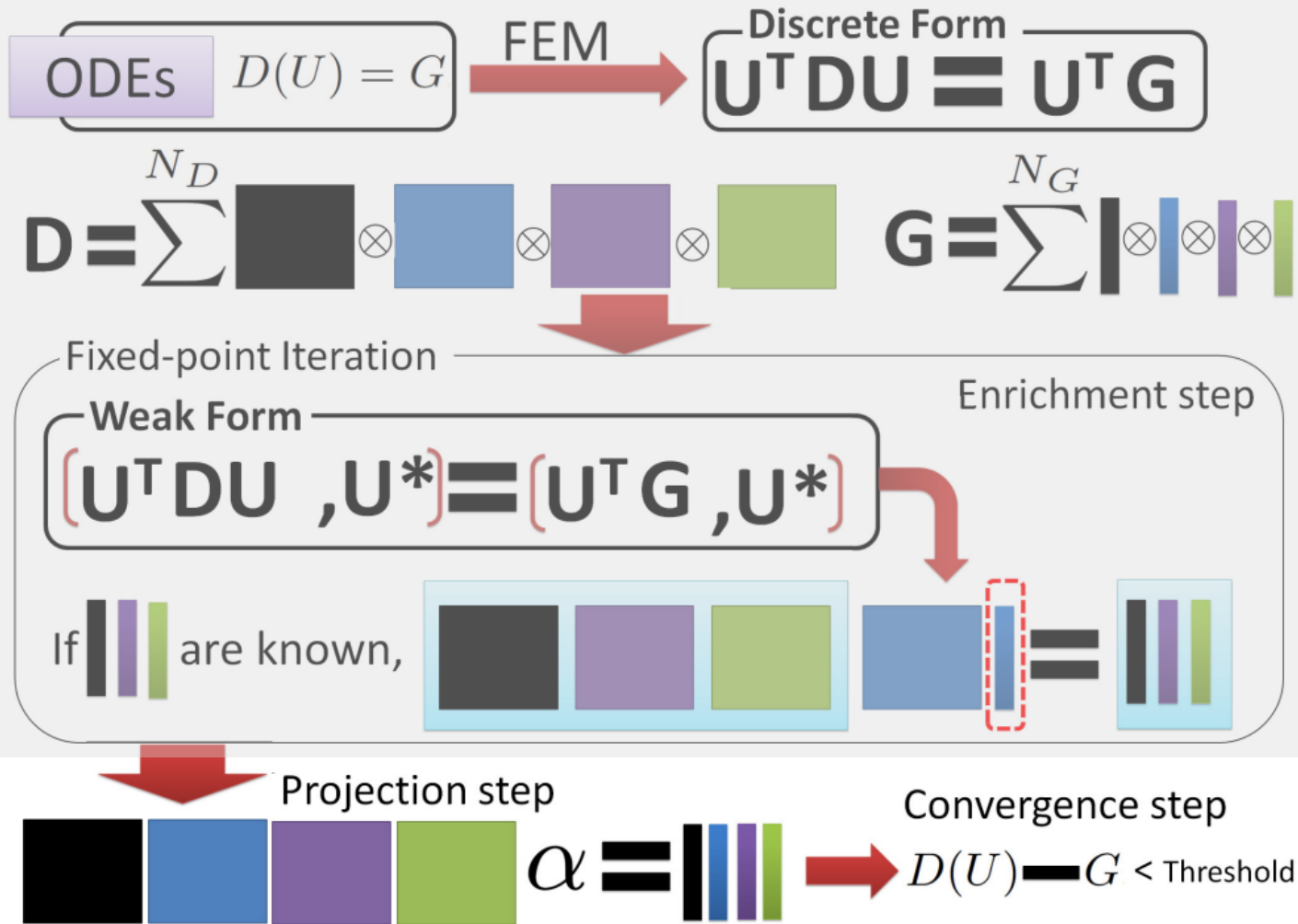
$$\|R_1^p \otimes R_2^p \dots \otimes R_d^p - R_1^{p-1} \otimes R_2^{p-1} \dots \otimes R_d^{p-1}\| < \epsilon$$



$$U = \sum_{i=1}^N \alpha_i U_1^i \otimes U_2^i \dots \otimes U_d^i \quad U_j^n = \frac{R_j}{\|R_j\|}$$

Projection Step

Model Reduction



Projection Step

Determine α_n

$$\langle D(U), \prod_{j=1}^d U_i^j(x_j) \rangle = \langle G, \prod_{j=1}^d U_i^j(x_j) \rangle \Rightarrow$$

$$\boxed{BA = H}$$

$$A = [\alpha_1 \alpha_2 \dots \alpha_n]^T$$

$$B_{ij} = \sum_{k=1}^{N_D} \left[\prod_{e=1}^d (U_e^i)^T D_e^k U_e^j \right]$$

$$H_i = \sum_{m=1}^{N_G} \left[\prod_{k=1}^d (U_k^i)^T G_k^m \right]$$

$$1 \leq i, j \leq n$$

Check convergence:

$$R^n = \sum_{i=1}^{N_D} \sum_{k=1}^n \alpha_i D_1^i U_1^k \otimes D_2^i U_2^k \dots \otimes D_d^i U_d^k - G < \epsilon$$

Coupled Model

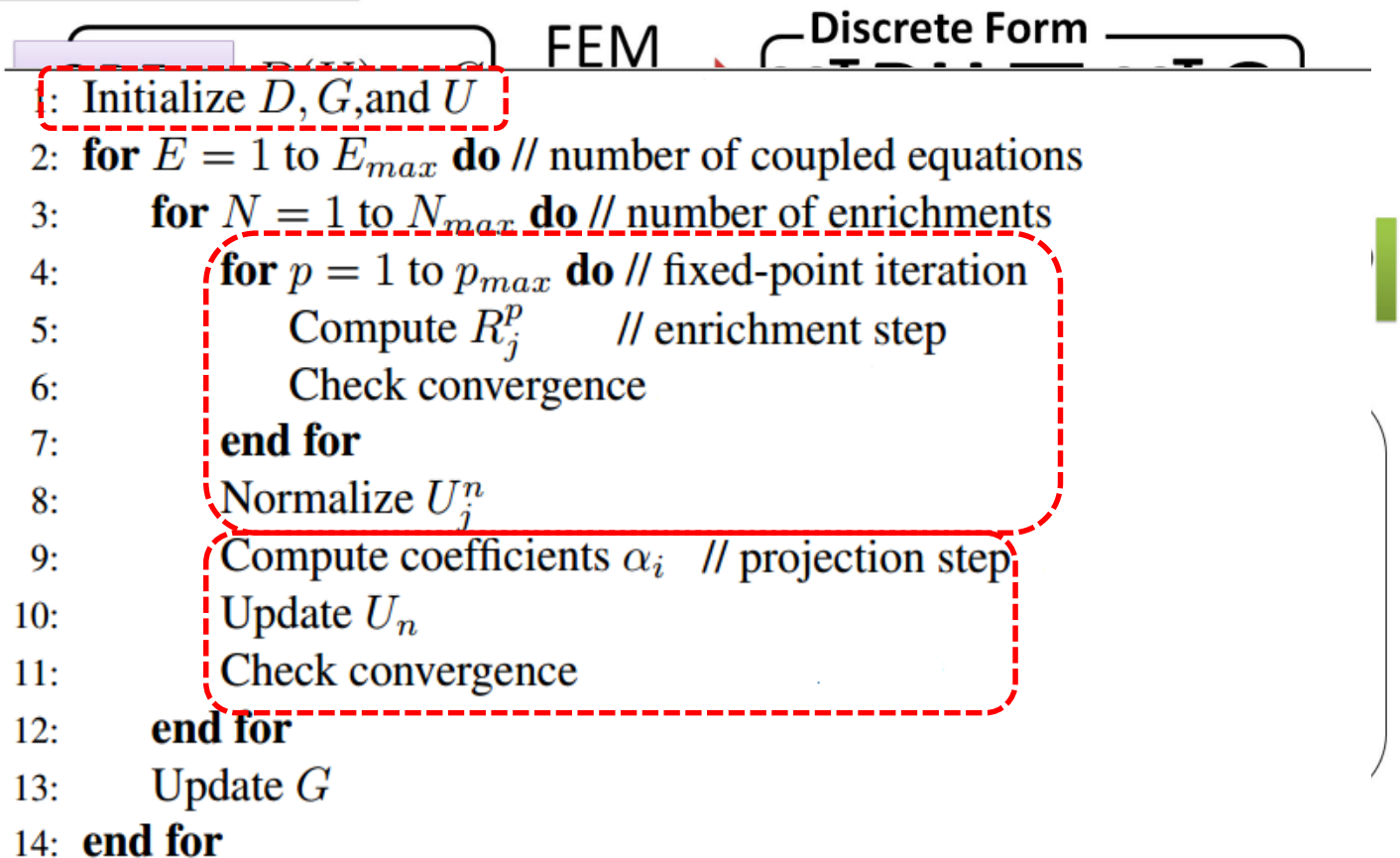
$$U = \sum_{i=1}^N \alpha_i U_1^i \otimes U_2^i \dots \otimes U_d^i + R_1 \otimes R_2 \dots \otimes R_d$$

$$W = \sum_{i=1}^N \beta_i W_1^i \otimes W_2^i \dots \otimes W_d^i + S_1 \otimes S_2 \dots \otimes S_d$$

$$\begin{cases} D_U(U, W) = G_U \\ D_W(U, W) = G_W \end{cases} \xrightarrow{\text{decoupling}} \begin{cases} D_U(U, W, R, S = 0) = G_U \\ D_W(U, W, R = 0, S) = G_W \end{cases}$$

A practical framework

Model Reduction



```

1: Initialize  $D, G$ , and  $U$ 
2: for  $E = 1$  to  $E_{max}$  do // number of coupled equations
3:   for  $N = 1$  to  $N_{max}$  do // number of enrichments
4:     for  $p = 1$  to  $p_{max}$  do // fixed-point iteration
5:       Compute  $R_j^p$  // enrichment step
6:       Check convergence
7:     end for
8:     Normalize  $U_j^n$ 
9:     Compute coefficients  $\alpha_i$  // projection step
10:    Update  $U_n$ 
11:    Check convergence
12:   end for
13:   Update  $G$ 
14: end for
  
```

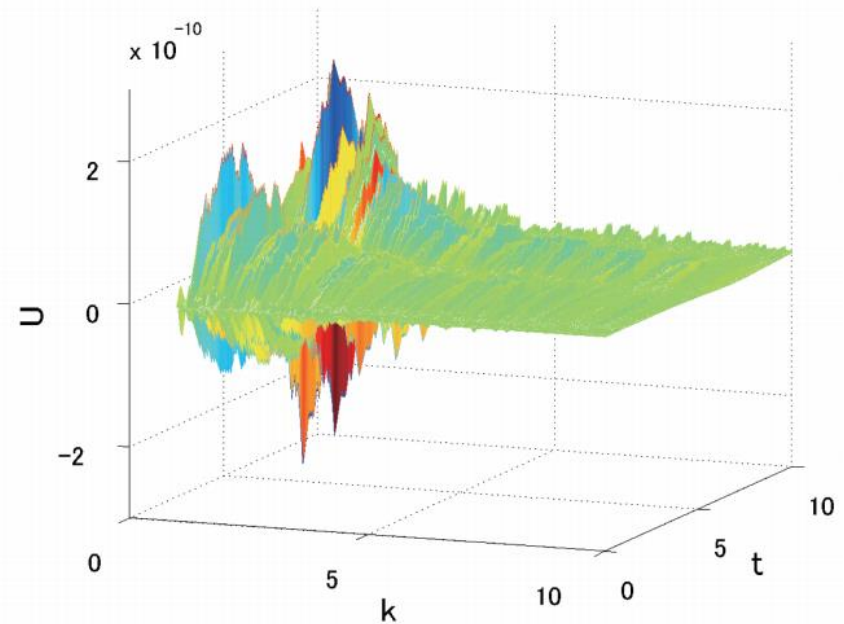
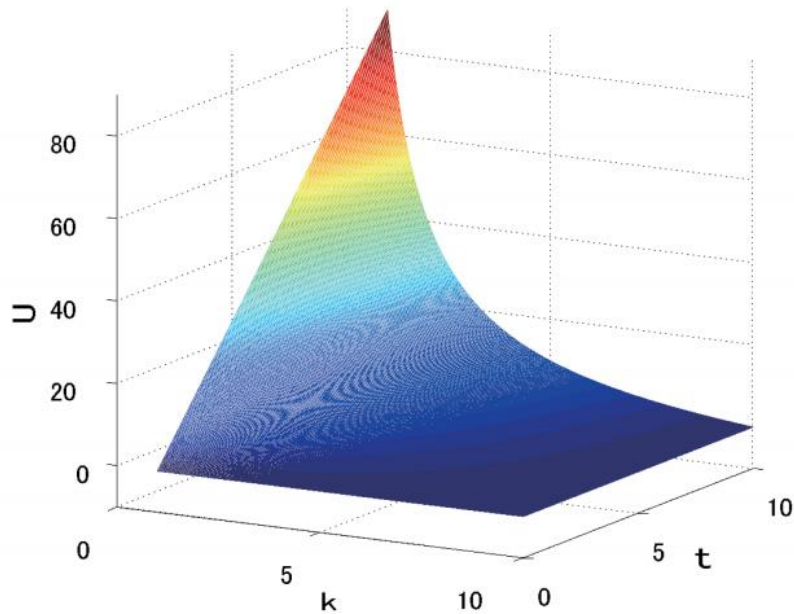

 $\alpha = \begin{bmatrix} \text{black} \\ \text{blue} \\ \text{purple} \\ \text{green} \end{bmatrix} \Rightarrow D(U) - G < \text{Threshold}$

RESULTS

Parametric model :Linear

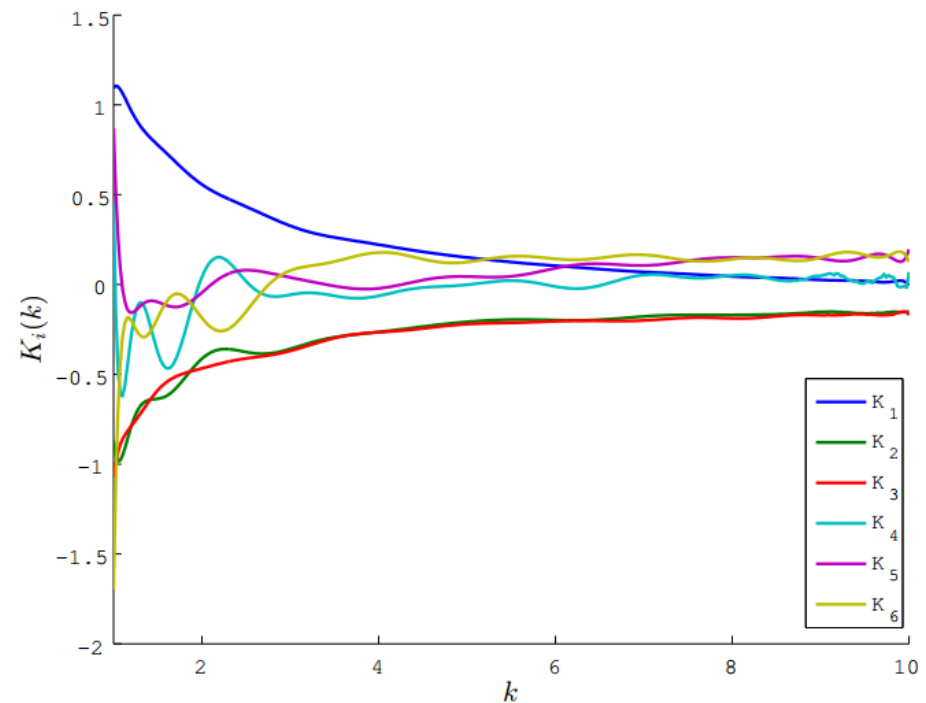
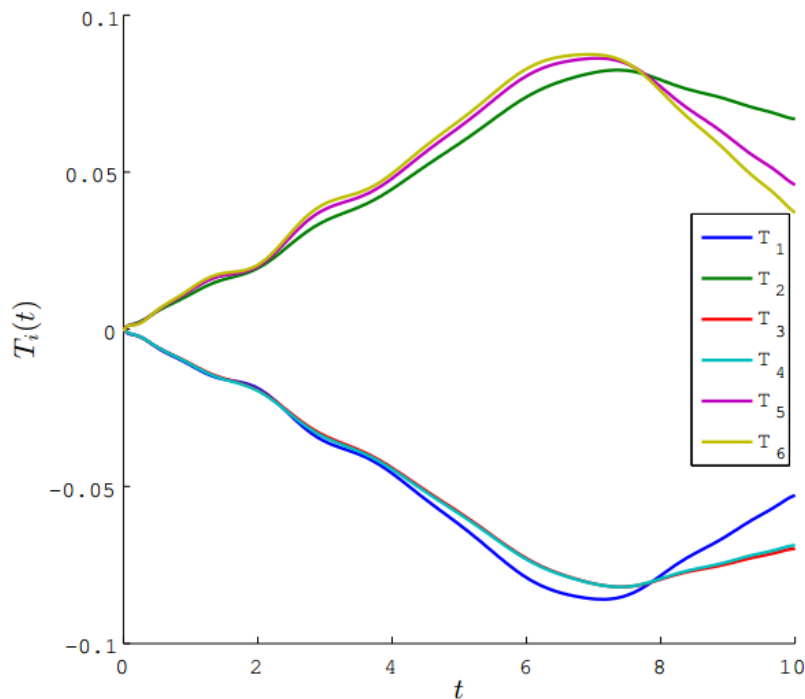
$$k\left(\frac{du}{dt} + 1\right) = 10$$
$$u(t=0) = 0$$

$$u(t, k) = \sum_{i=1}^n \alpha_i T_i(t) K_i(k)$$



Parametric model: Linear

- Base functions

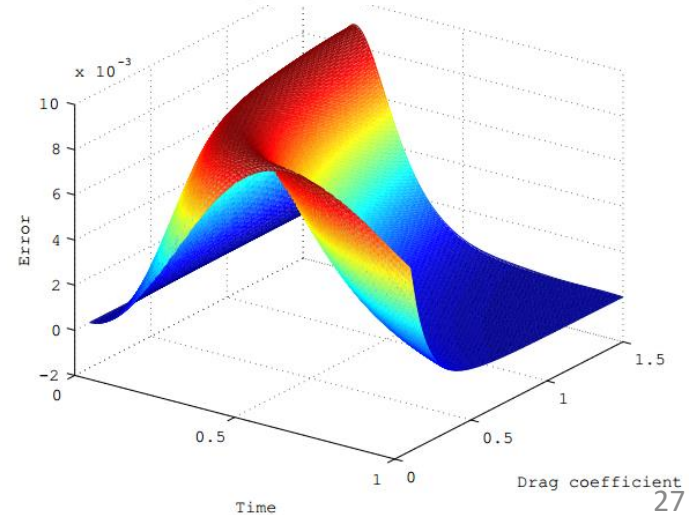
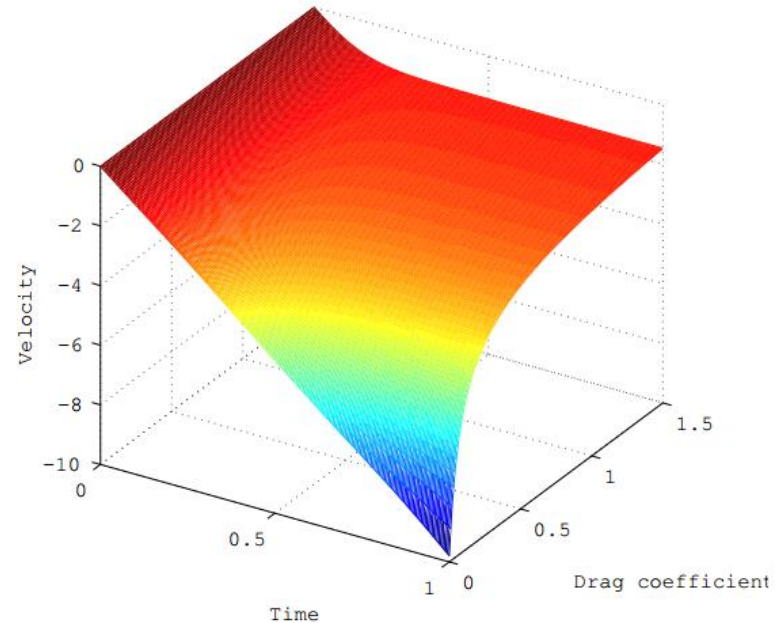
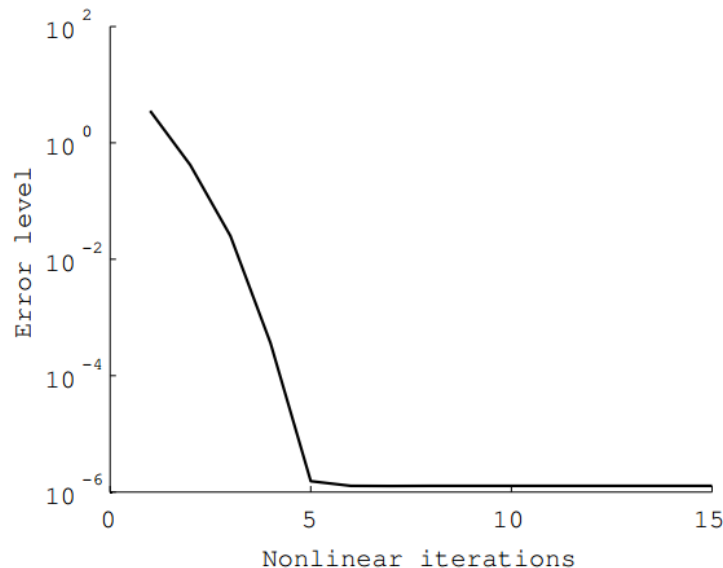


Parametric model :Nonlinear

1D free-fall problem

$$\frac{du}{dt} = \frac{1}{2m} \rho_f C_D A u^2 - g$$

$$u(t, C_D) = \sum_{j=1}^n \alpha_j T_j(t) D_j(C_D)$$

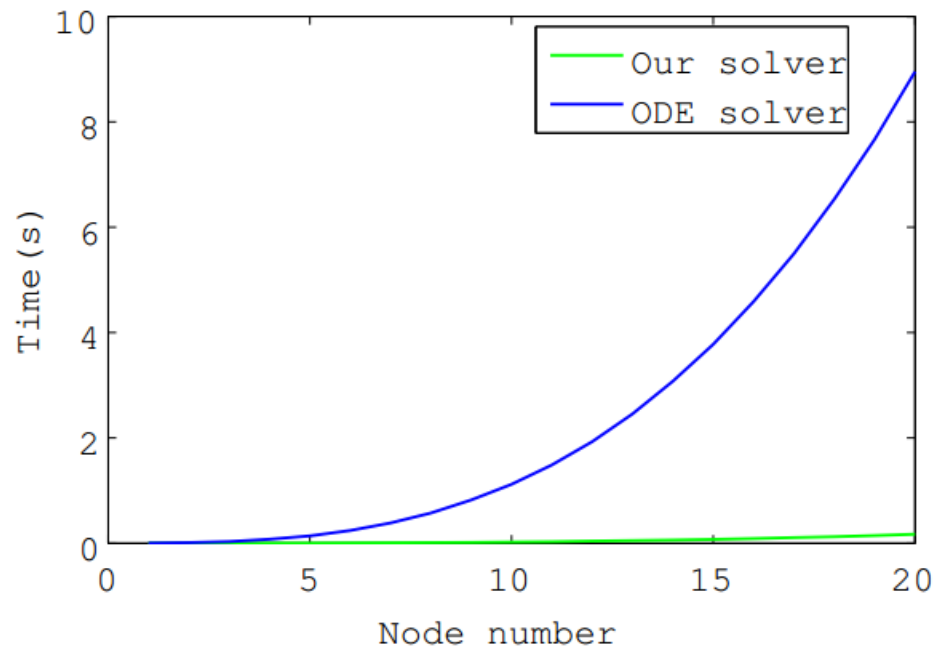
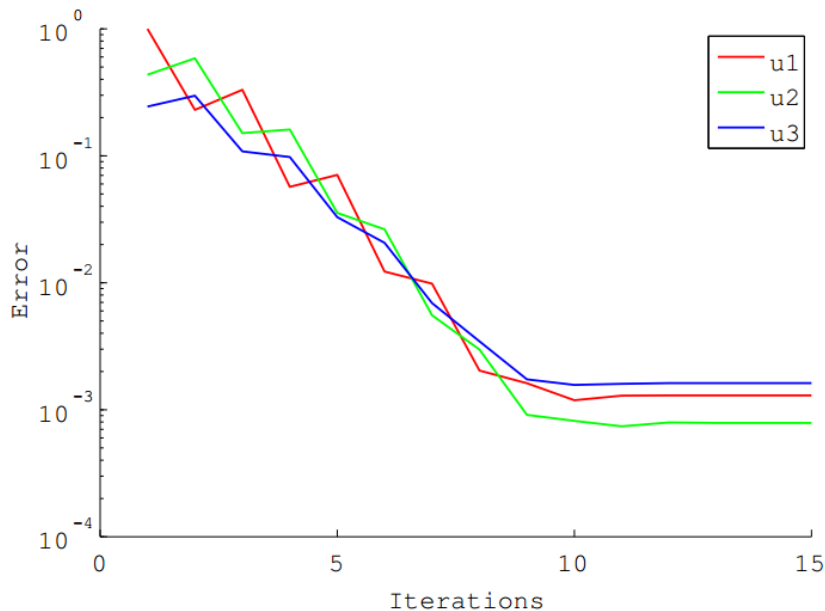


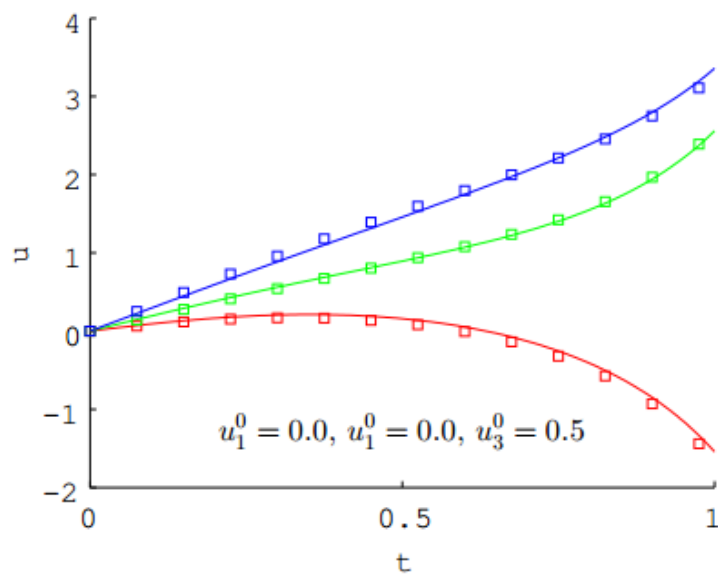
Coupled Model

$$\begin{cases} \frac{du_1}{dt} + u_2 u_3 = 1 \\ \frac{du_2}{dt} + u_1 u_3 = 2 \\ \frac{du_3}{dt} + u_1 u_2 = 3 \end{cases}$$

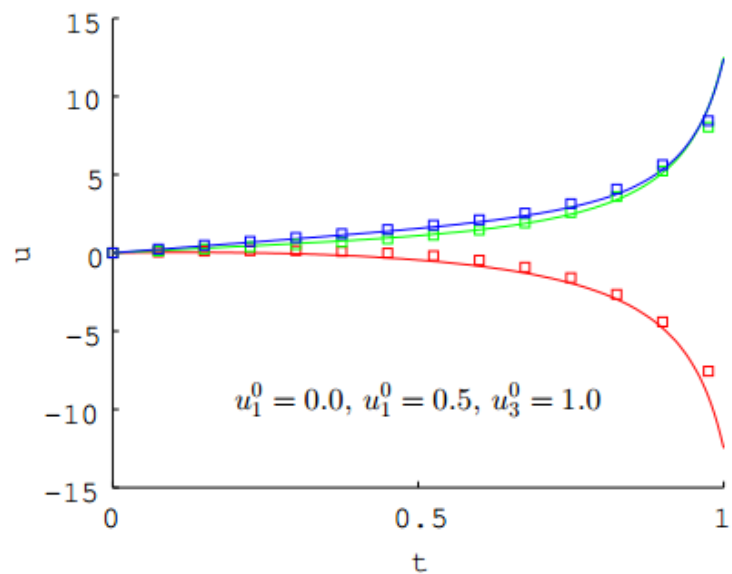
$$\hat{u}_i(t, u_i^0) = \sum_{j=1}^n \alpha_j T_j(t) U_j^1(u_1^0) U_j^2(u_2^0) U_j^3(u_3^0)$$

$$u_i(t = 0) = u_i^0, i = 1, 2, 3$$

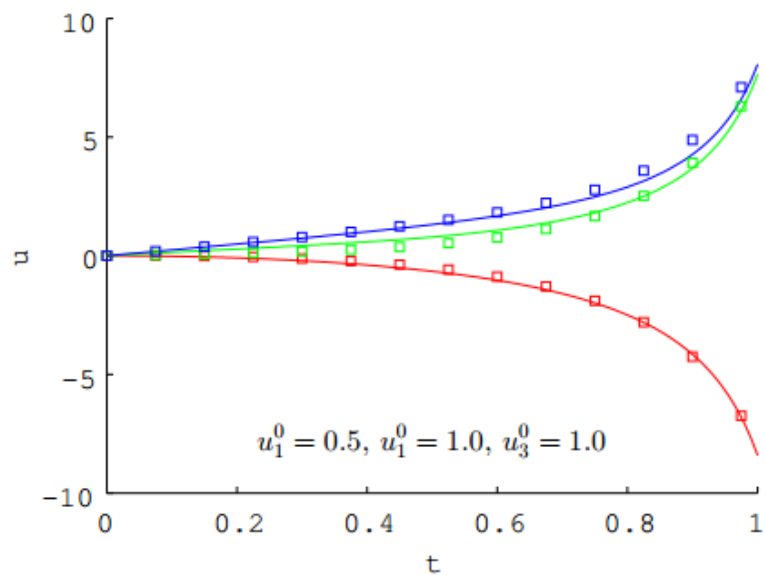
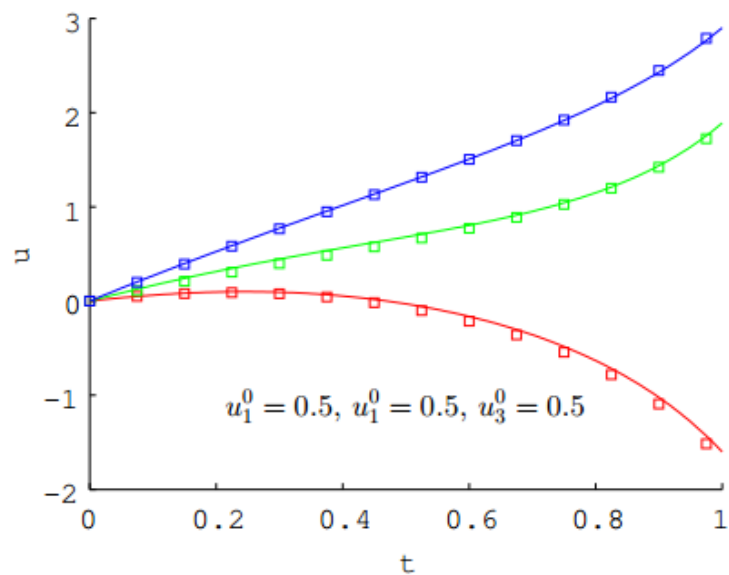




(a)



(b)



Coupled Model: Stiff

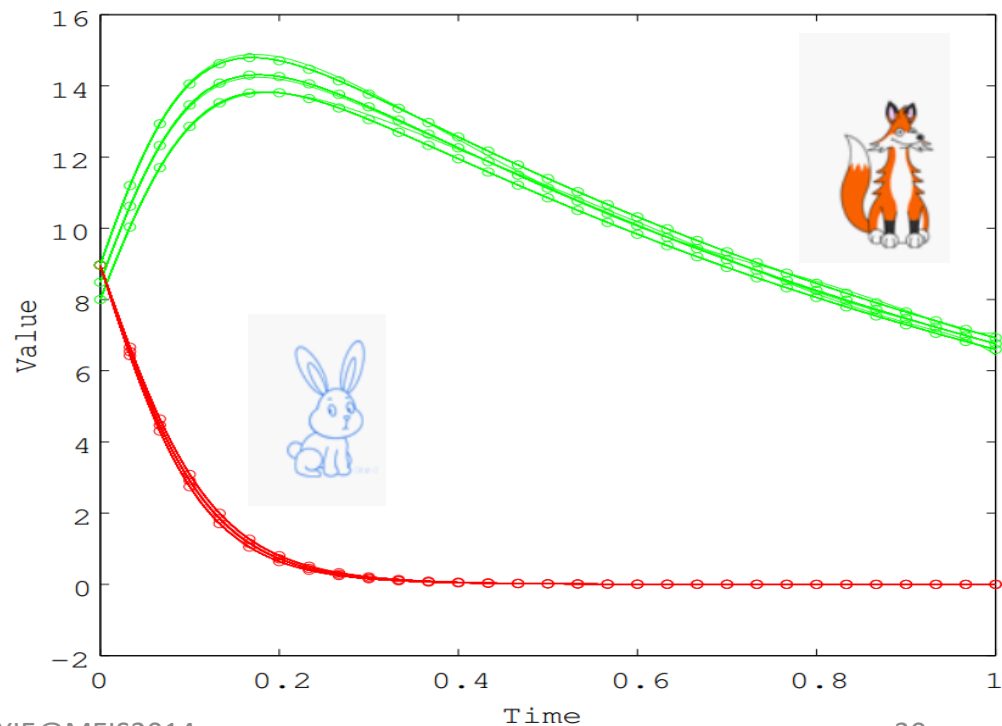
Prey–Predator model

Lotka–Volterra equations:

$$x' = x(a - by)$$

$$y' = -y(c - dx)$$

$$x(t, x_0, y_0) = \sum_{i=1}^n \alpha_i T_i(t) X_i(x_0) Y_i(y_0)$$
$$y(t, x_0, y_0) = \sum_{j=1}^n \alpha_j T_j(t) X_j(x_0) Y_j(y_0)$$



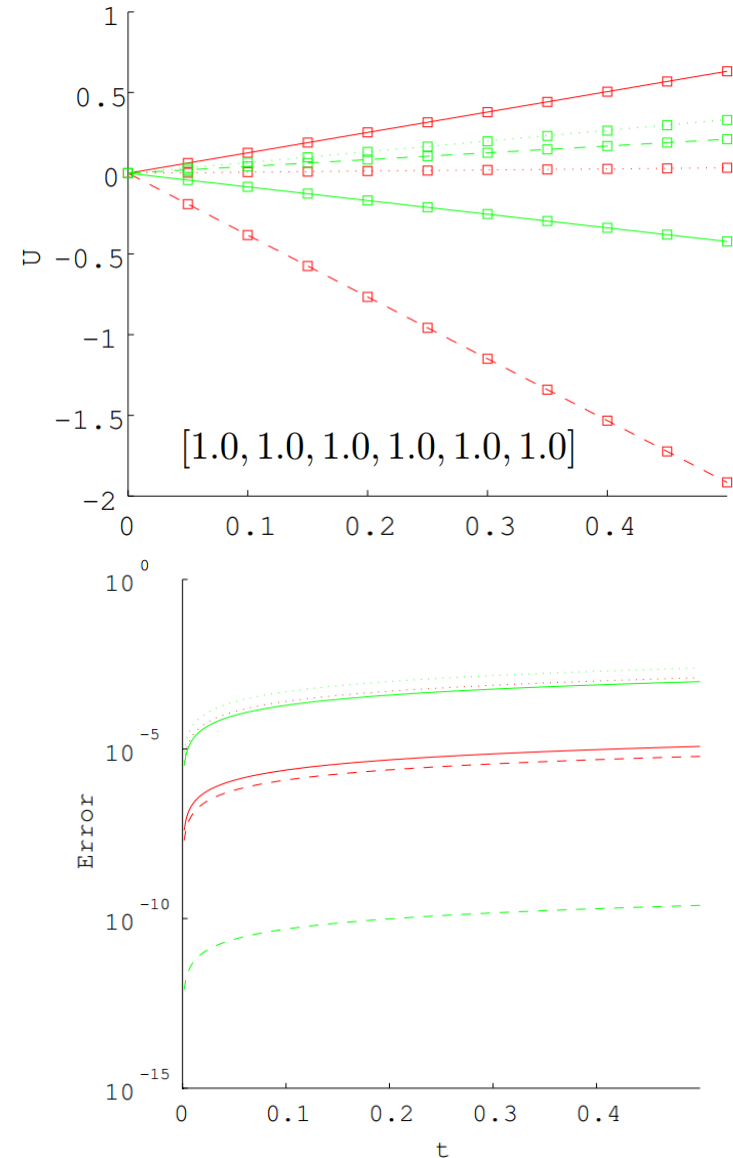
Coupled Model: Complex

Underwater rigid body dynamics
(without velocity coupling)

$$\begin{cases} (mE + M) \frac{du}{dt} = (mE + M)u \times \omega + f_g \\ (J + I) \frac{d\omega}{dt} = (J + I)\omega \times \omega + (Mu) \times u + \tau_g \end{cases}$$

$$U_k(t, u_k^0) = \sum_{i=1}^n \alpha_i T_i(t) \prod_{j=1}^6 U_i^j(u_j^0)$$

Red: Translational velocity
Green: Angular velocity

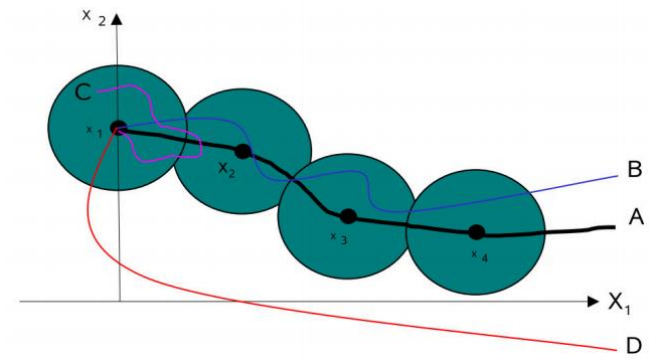


Conclusion

- A practical framework of separated representation
- A promising model-driven approach for computer animation
- Reduced Model for weakly-coupled and nonlinear models

Limitation and Future work

- ◆ Graphics application in progress
- ◆ Strongly coupled and nonlinear model
 - ◆ Nonlinear Model Reduction (Trajectory PieceWise-Linear approach (TPWL), Discrete Empirical Interpolation Method)
 - ◆ Hybrid reduced approach with **POD**



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THANK YOU